

# Lecture 20

Plan: 1) ~~Finish LCIS algo.~~  
2) Matroid intersection polytope  
3) Start min-cost arborescent  
after today, ~4 more lectures! ~2 matroids  
~3 ellipsoid

## Matroid intersection polytope

- Let  $M_1 = (E, I_1)$  &  $M_2 = (E, I_2)$  be two matroids, rank functions  $r_1, r_2$ .

- Analogously to the matroid polytope, let

$$X = \left\{ \begin{matrix} 1_S \in \{0,1\}^E \\ \in \mathbb{R}^E \end{matrix} : S \in I_1 \cap I_2 \right\}$$

i.e.  $X \subseteq \mathbb{R}^E$  is set of indicators of common independent sets.

- Define the matroid intersection

polytope  $P_{M_1, M_2} := \text{conv}(X)$

(can use to optimize linear functions over  $X$ ).

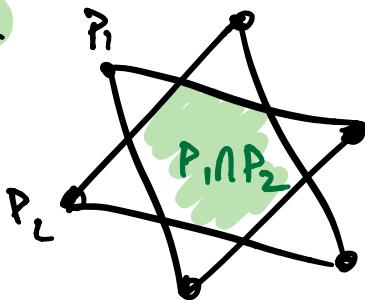
- Main result:  $P_{M_1, M_2}$  is the intersection  $P_{M_1} \cap P_{M_2}$  of the matroid polytopes  $P_{M_1}, P_{M_2}$  of  $M_1, M_2$ .

$$\Rightarrow \text{vertices}(P_{M_1}) \cap \text{vertices}(P_{M_2}) = \text{vertices}(P_{M_1} \cap P_{M_2})$$

- This is surprising! In general, for polytopes  $P_1, P_2$   
 $\text{verts}(P_1) \cap \text{verts}(P_2) \neq \text{verts}(P_1 \cap P_2)$

$\Rightarrow \text{vertices}(P_{M_1}) = \text{vertices of } I \Rightarrow$  intersection is common indep sets.  
= vertices of  $P_{M_1, M_2}$   
if  $P_{M_1, M_2} = P_{M_1} \cap P_{M_2}$ , then same set of vertices.

e.g.



$P_1, P_2$  share no vertices but

$P_1 \cap P_2 \neq \emptyset$   
(& hence has vertices).

- In terms of inequalities?
- Recall matroid polytope:  
for  $r$  rank function of  $M$ ,

$$P_M = \left\{ x \in \mathbb{R}^E : x(S) \leq r(S) \quad \forall S \subseteq E \right. \\ \left. x_e \geq 0 \quad \forall e \in E \right\}.$$

$\vdash \sum_{e \in S} x_e.$

- $P_{M_1} \cap P_{M_2}$  has both sets of constraints, so  
1981: can efficiently decide membership in  $P_{M_1, M_2}$ .

Theorem:

Let  $P = P_M_1 \cap P_{M_2}$ , i.e.

$$P = \left\{ x \in \mathbb{R}^E : \begin{array}{l} x(S) \subseteq f_1(S) \quad \forall S \subseteq E \\ x(S) \subseteq f_2(S) \quad \forall S \subseteq E \\ x_e \geq 0 \quad \forall e \in E \end{array} \right\}$$

Then

$$P_{M_1, M_2} = P \quad \text{i.e. } P_{M_1, M_2} = P_{M_1} \cap P_{M_2}.$$

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Proof: Plan: similar to lecture 17,  
vertex proof for matroid  
polytope.

- Like second proof for matroid polytope, use vertex integrality
- Integrality <sup>of P</sup> suffices by the usual logic:
  - ▷ Clearly  $\text{conv}(X) \subseteq P$ , b/c  $X \subseteq P_{M_1} \cap P_{M_2} = P$ .

▷ On the other hand, if  $P$  integral then  
 $P \subseteq \text{conv}(X)$  because integral  
points in  $P_{M_1}, P_{M_2}$  are indicators of  
indep sets in  $M_1, M_2 \Rightarrow$  integral  
points in  $P_{M_1} \cap P_{M_2} = P$  are common indep sets.

- Again, if  $P = \{x : Ax \leq b, x \geq 0\}$ ,  
matrix  $A$  is not totally unimodular.

there are matrices

- But submatrices describing vertices  
will be T.U. (e.g.).



$$x^* \in P \quad x^* = (1, 1, \sqrt{2})$$

Let  $x^*$  be an extreme point of  $P$ .  
↳ i.e. vertex

- We know  $x^*$  characterized  
by which inequalities are tight  
for it.

- For  $i \in \{1, 2\}$ , let  $\bigcup_{e \in S} x^* e = x^* \cdot \mathbb{I}_S$ .

$$T_i = \left\{ S \subseteq E : x^*(S) = r_i(S) \right\}$$

i.e.  $T_i$  sets of tight rank constraints in  $M_i$ .

- Let  $J = \{e : x^* e = 0\}$ .

- Then  $x^*$  is unique solution to

$$F_1 \left[ \begin{array}{ll} x(S) = r_1(S) & \forall S \in T_1 \\ x(S) = r_2(S) & \forall S \in T_2 \\ x_e = 0 & \forall e \in J. \end{array} \right]$$

- That is,  $\{x^*\}$  is the intersection of two faces  $F_1, F_2$  in  $P_{M_1}, P_{M_2}$ .

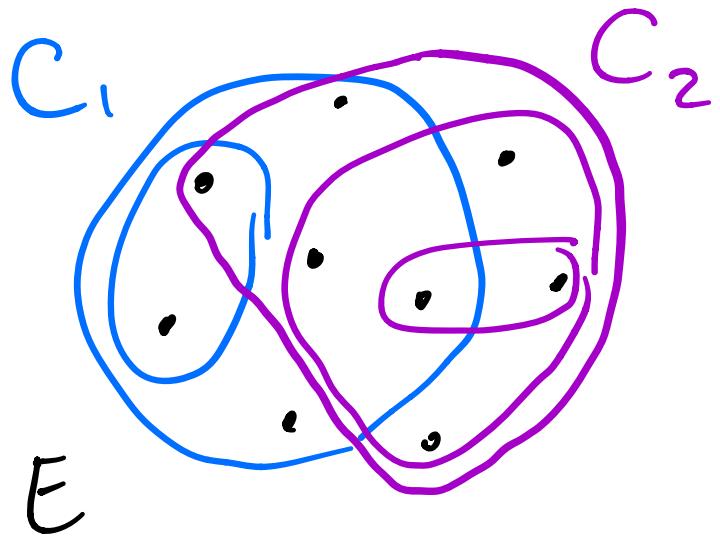
$$F_i = \{x \in P_{M_i} : x(s) = r_i(s) \ \forall s \in T; x_e = 0 \ \forall e \in J\}.$$

- Recall from lecture 17:  $T$  can be replaced by a chain  $C_i$  without changing  $F_i$ .

$\exists C_1, C_2$  chains s.t.

$$F_i = \{x \in P_{M_i} : x(s) = r_i(s) \ \forall s \in C_i; x_e = 0 \ \forall e \in J\}.$$

e.g.



• Thus, assume  $x^*$  is solution to

$$x(s) = r_1(s) \quad \forall s \in C_1$$

$$x(s) = r_2(s) \quad \forall s \in C_2$$

$$x_0 = 0 \quad \forall t \in J.$$

- This is  $Ax \approx b$  for  $b \in \mathbb{R}$

**Claim:** A T.U.

$\Rightarrow x^*$  integral.

- Why? Rows of  $A$  are 1s of  $S$  in chain  $C_1$  or  $C_2$ .

e.g.

$A =$   
can permute rows  
of  $A$  so it  
looks like this.  
doesn't change T.U.

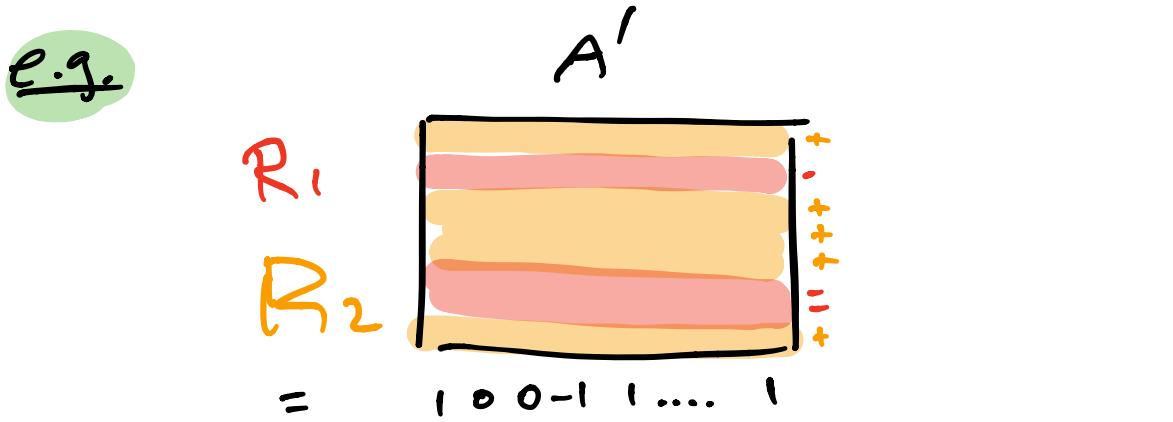
$$\left[ \begin{array}{ccccc} 1 & 1 & 1 & 1 & \\ 0 & 1 & 1 & 1 & \\ 0 & 0 & 0 & 1 & \\ 1 & 1 & 0 & 1 & \\ 1 & 1 & 0 & 0 & \\ 0 & 1 & 0 & 0 & \end{array} \right] \quad \left. \begin{array}{l} \{ C_1 \\ \{ C_2 \end{array} \right]$$

- We use discrepancy to prove.  
Theorem 3.4 in polyhedral notes

- Recall: A T.U.  $\Leftrightarrow$   $\Delta$  submatrices

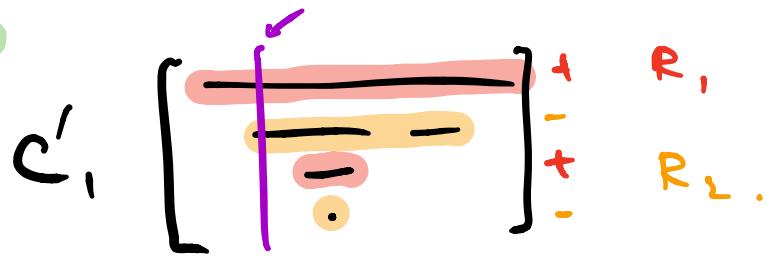
$A'$  of  $A$ ,  $\exists$  partition  $R_1, R_2$  of rows of  $A'$   
 $\uparrow$   
 $a_i$

$\sum_{i \in R_1} a_i - \sum_{i \in R_2} a_i$  has  $\{-1, 0, +1\}$ .



- Consider submatrix  $A'$  of  $A$  corresponds to subchains  $C'_1, C'_2$  (same form as  $A$ )
- Assign  $R_1, R_2$  as follows:
  - ▷ Assign target elmt of  $C'_1$  to  $R_1$ , then alternately assign remaining elts of  $C'_1$  to  $R_2, R_1$

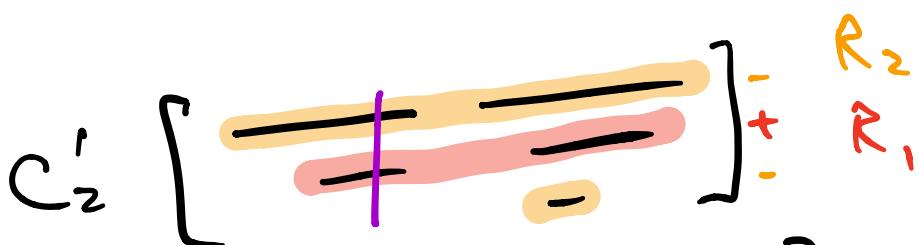
e.g.



sum has entries in  $\{0, 1\}$ .

► For  $C'_2$ , assign oppositely.

e.g.



sum has entries in  $\{0, -1\}$ .

- Overall, sum has entries in  $\{-1, 0, 1\}$ .
- completes the proof.

□.

Matroid intersection  
optimization

- Given a cost function  $c: E \rightarrow \mathbb{R}$   
Can we efficiently compute

$$\max_{S \in I_1 \cap I_2} c(S) := \sum_{e \in S} c(e) = c \cdot \mathbf{1}_S.$$

equiv: optimize  $C^T x$  over  $x \in P_{M_1, M_2}$ .

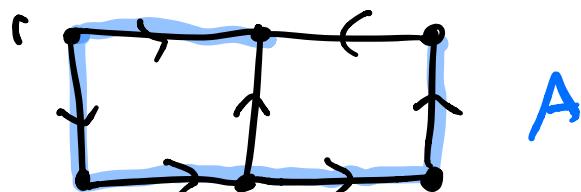
- For just one matroid: greedy alg works.
- For  $c = \mathbf{1}_{(1, \dots, 1)}$ : just L.C.I.S.  
 $(1, \dots, 1)$  1E1 times
- For perfect matching: Hungarian alg.  
e.g. min-cost p.m.
- Can also compute min cost L.C.I.S.  
Exercise: equiv. to max cost indep for  $c' = k - c$  for  $k$  large.

- In general, YES, can efficiently compute.
  - ▷ ellipsoid
  - ▷ complicated primal-dual algos.
    - strongly poly. time
    - & steps indep of  $C$
    - if arithmetic is unit cost.
- ~~Today!~~: simpler primal dual alg. for

## Min-Cost arborescence

- Recall: given directed graph  $D$  & vertex  $r$ , arborescence  $A$  is a spanning tree in  $D$  directed away from  $r$ .

e.g.:



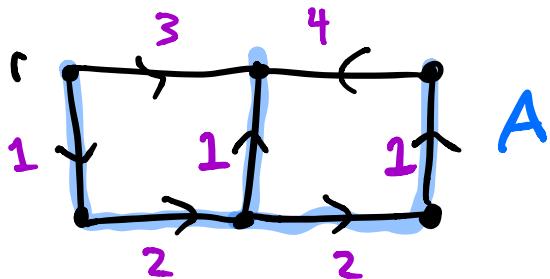
- min-cost arborescence:

$$\min \sum_{e \in A} c(e) = c(A)$$

at arborescence  $e \in A$

e.g.

$c$



$A$

- e.g. edges = roads to be fixed

$r$  = distribution center

Cost = expense of fixing road.

- First, I.P. formulation:

assume  $c$  nonnegative.

think  $x = 1_A$   
for  $A$  arborescent  
 $\sum c_e x_e = c(A)$

$$OPT = \min_{x \in \mathbb{R}^E} \sum_{e \in E} c_e x_e$$

subject to

$x = \sum_{e \in \delta^-(s)} x_e$ , where  
A has no cuts.



$$\sum_{e \in \delta^-(s)} x_e \geq 1 \quad \forall s \in V - r$$

$$\sum_{e \in \delta^-(v)} x_e = 1 \quad \forall v \in V - r$$

indegree = 1  
except r.

$$x_e \in \{0, 1\}$$

$x$  is an  
indicator.

- Check: only solutions are  $\mathbb{1}_A$

where A is an arborescence.

in particular, all arbos satisfying constraints.

- Miraculously, we'll show even w/out integrality constraint & indegree constraint, there's still an optimal solution that's an arborescence.

- I.e. the following L.P. has optimizer  
 $\mathbf{1}_A$  st.  $A$  is an arborescence.

$$LP = \min_{x \in R^E} \sum_{e \in E} c(e)x_e$$

subject to  $\sum_{e \in \delta(s)} x_e \geq 1 \quad \forall s \in V - r$ .

(primal)  $x_e \geq 0 \quad \forall e \in E$

(note  $LP \leq OPT$  b/c LP has fewer constraints).

(typo in pre-lecture!)

$\xrightarrow{\text{"symmetric version"}}$

- Dual LP is

$$LP = \max \sum_{S \subseteq V-r} y_S$$

subject to

$$\begin{aligned} & \sum_{S: e \in \delta^-(S)} y_S \leq c(e) \quad \forall e \in E \\ (\text{dual}) \quad & y_S \geq 0 \quad \forall S \subseteq V-r. \end{aligned}$$

- Algorithm sketch: construct

▷ arb.  $A$ ,  
 ▷ dual. fees.  $y$   
 satisfying complementary slackness

Then  $c(A) = LP$ , but  $LP \leq OPT$

$c(A) \leq OPT \Rightarrow \boxed{c(A) = OPT.}$

- Complementary slackness for  $x = 1_A, y$  says:
  - a.)  $y_s > 0 \Rightarrow |A \cap \delta^-(s)| = 1$
  - b.)  $e \in A \Rightarrow \sum_{s: e \in \delta^-(s)} y_s = c(e).$
- Two phases of algorithm:
  - 1) Construct
    - ▷ dual feas  $y$
    - ▷ set  $F$  of edges s.t.  
every vertex of  $\cup \tau_j$   
reachable from  $\Gamma$  in  $F_j$

$F$  might not be an arborescence.

&  $y, x = F$  satisfies (b).

2) Remove unnecessary edges from  $F$ , get arborescence which satisfies both (a) & (b).

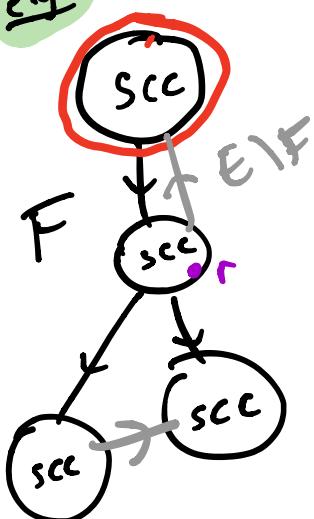
## Phase 1

Initialize  $F = \emptyset$ ,  $y = 0$   
counter  $K = 1$

▷ While not everything reachable from  $r \in F$

▷ select  $S \subseteq V - r$

e.g.



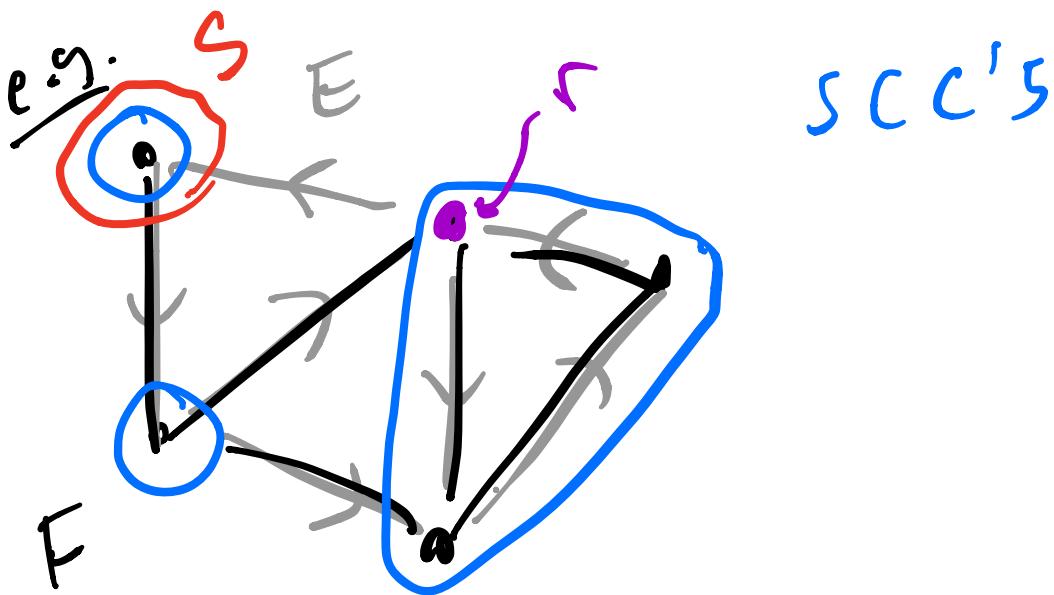
i)  $F$  strongly connected in  $S$

(every vertex can reach every other using only edges contained entirely in  $S$ )

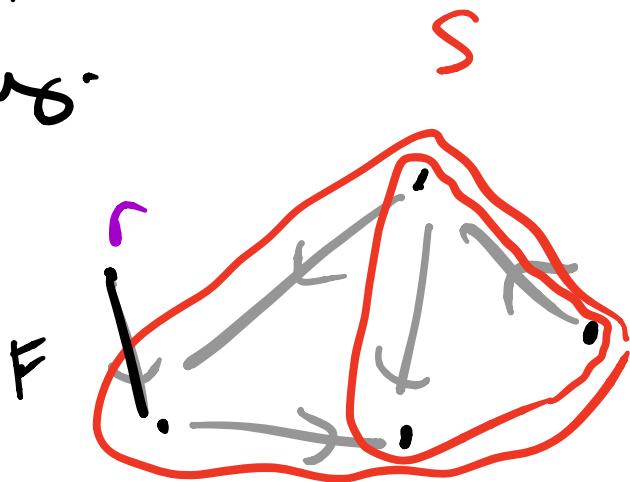
ii)  $F \cap \delta^-(S) = \emptyset$

$S$  is a "source" in  
decomp. of  $F$  into S.C.C.'s.

(digraph has decomp where if S.C.C.'s are contracted,  
left with DAG).



$S$  is a subset of vertices,  
 $F$  subset of edges.  
 does  $S \subseteq$  vertices "touched by  $F$ "?  
 not necessarily.  
 initially



▷ increase  $y_S$  until new  
inequality

$$\sum y_S \leq c(e_k)$$

$S : e_k \in \delta(S)$   
becomes an equality. note  $e_k \notin F$

( $y$  remain dual feas,  
b/c it was before)

$\forall e \in F \cap \delta(S) = \emptyset.$

▷  $F \leftarrow F + e_k, k \leftarrow k+1$

new  $F, y$  don't violate (b)  
because  $e_k$  is tight.

▷ Return  $F, y$  satisfying (b),  
& every  $y$  reachable from  
 $r$  in  $F$ .

Phase 2: eliminate as many edges as we can in reverse order they were added.

▷ For  $i=k \dots 1$ :

▷ If  $F - e_i$  contains a directed path from  $r$  to every vertex,  $F \leftarrow F - e_i$ .

▷ Return  $A = F$

Claim 1:  $A$  is an arborescence

Pf: We'll show  $|A| = N - 1$  &  $d^-(v) = 1 \forall v \in V - r$

- If indegree  $< 1$  for  $v \neq r$ , contradicts reachability in  $A$
- if  $|A| > |V| - 1$  then collision  $e_i \rightarrow e_j$

- suppose  $i < j \Rightarrow$  in reverse delete, would have removed  $e_j$ . b/c any vertex reached through  $e_j$  is reachable through  $e_i$ .  $\square$

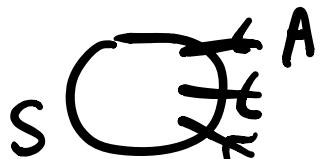
finally:

Claim 2: Condition (a) of complementary slackness holds.

a.)  $y_s > 0 \Rightarrow |A \cap \delta^-(s)| = 1$ .

PF: Assume not  $\exists S$  s.t.

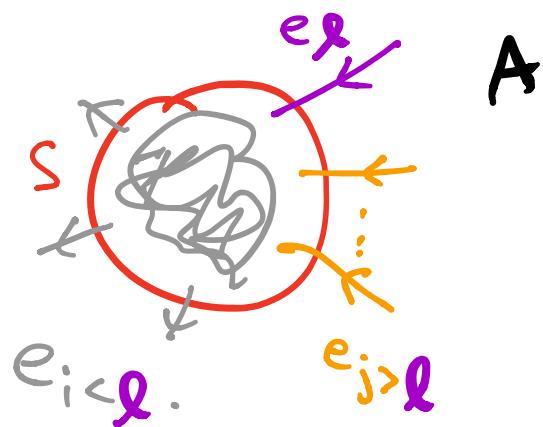
$$y_s > 0 \text{ & } |A \cap \delta^-(s)| > 1$$



- $S$  was chosen at some step  $t$  of phase I when we added  $e_S$  to  $F$ .

- $F$  had no other edges in  $\delta^-(s)$  when  $e_\ell$  was added.  
 $(\text{by construction})$   
 $\Rightarrow$  all edges of  $A \cap \delta^-(s)$   
are  $e_j$  for  $j > \ell$ .

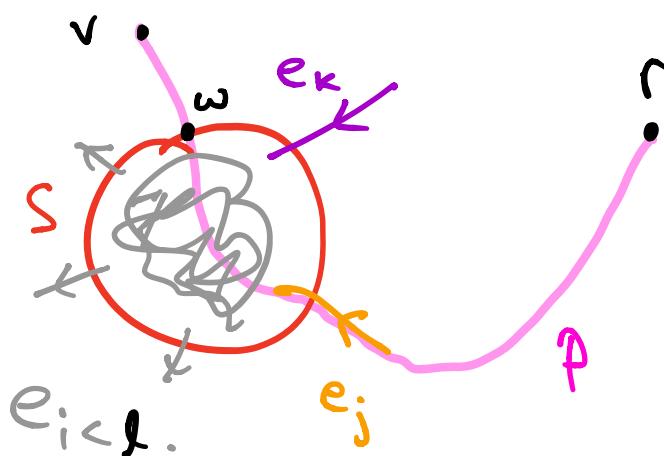
- When  $S$  chosen,  $F$  strongly connected within  $S$   
 $\Rightarrow S$  strongly connected using  
only  $e_i$   $i < \ell$ .



- Subclaim: All  $e_j$ ,  $j > l$  should have been removed in Phase 2.

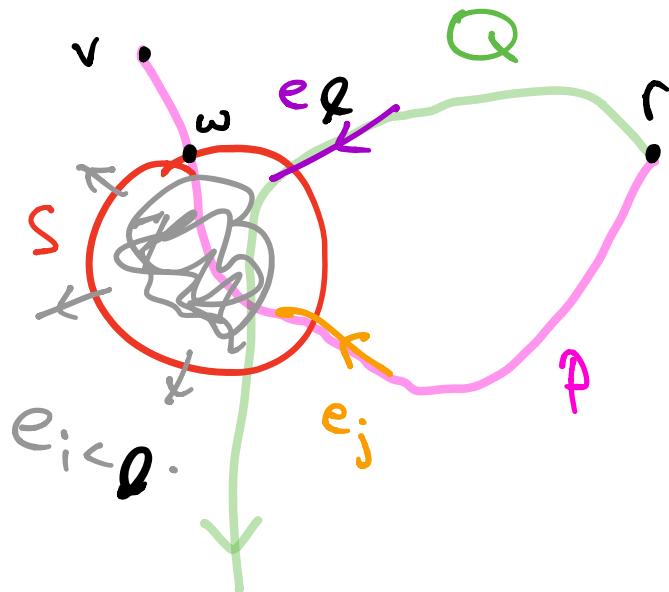
Why? suppose  $e_j$  necessary to visit some vertex  $v$

- let  $P: r \rightarrow v$  path using  $e_j$
- Let  $w$  last vertex in  $S$  on  $P$



Note:  $P$  first enters  $S$  through  $e_j$ , else could short out  $e_j$  because  $S$  stop conn.

- Because  $e_2$  is necessary at Step l of Phase 2, there must be another path Q through  $e_k$ .



similarly: Q must enter S first than  $e_l$ .

- Can use Q to shortcut  $e_j$ ; thus  $e_j$  not necessary.

